

Mixture models for phylogenetic analysis in IQ-TREE2

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Algorithmic Advances and Implementation Challenges:
Developing Practical Tools for Phylogenetic Inference (Nov 18 - 22, 2024)



Australian
National
University

Model-based Phylogenetics

DNA/Amino acid Sequence Alignment

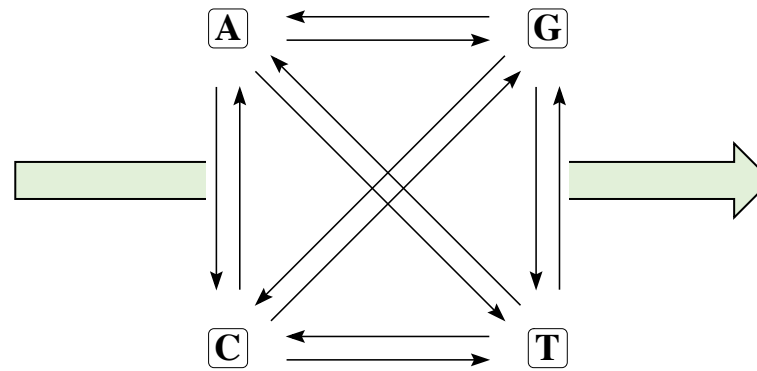
Bird	C	A	A	-	-	-	A	A	T	A
Crocodile	C	A	C	A	-	T	A	C	-	T
Turtle	C	A	C	T	A	T	A	A	G	T
Human	C	A	C	-	-	-	A	C	A	A

Assumptions are:

1. Single model
2. Single tree
3. Sites evolve independently
4.

Markov model

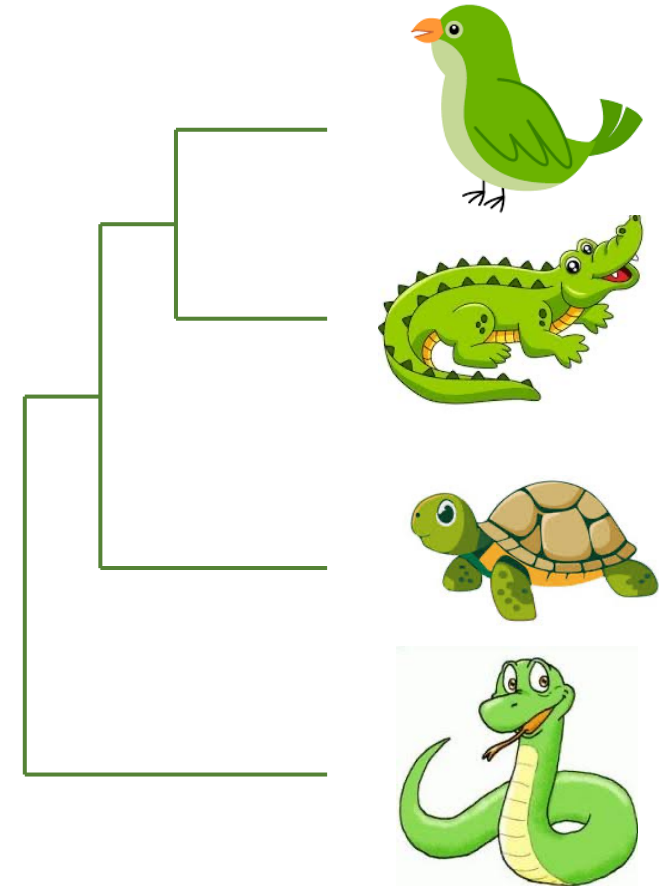
Substitution matrix:



Frequency array:

$$\pi_A \quad \pi_C \quad \pi_G \quad \pi_T$$

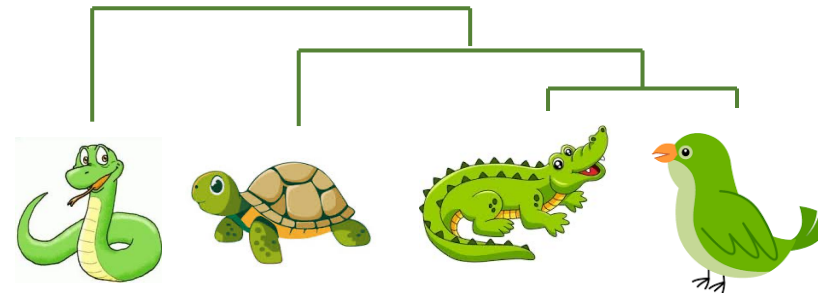
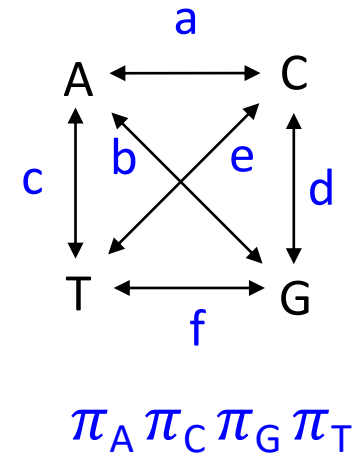
Phylogenetic Tree



Assumption

All alignment sites evolve equally under the same phylogenetic process?

Bird	C	A	A	-	-	-	A	A	T	A
Crocodile	C	A	C	A	-	T	A	C	-	T
Turtle	C	A	C	T	A	T	A	A	G	T
Human	C	A	C	-	-	-	A	C	A	A



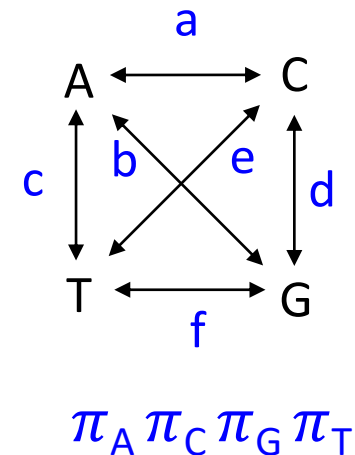
Single model often not true

For example:

- Multiple genes
- On different chromosomes
- Encode different proteins
- Proteins have very different functions
- Under different biological constraints

➤ One model is not good enough to describe the evolutionary process along the alignment

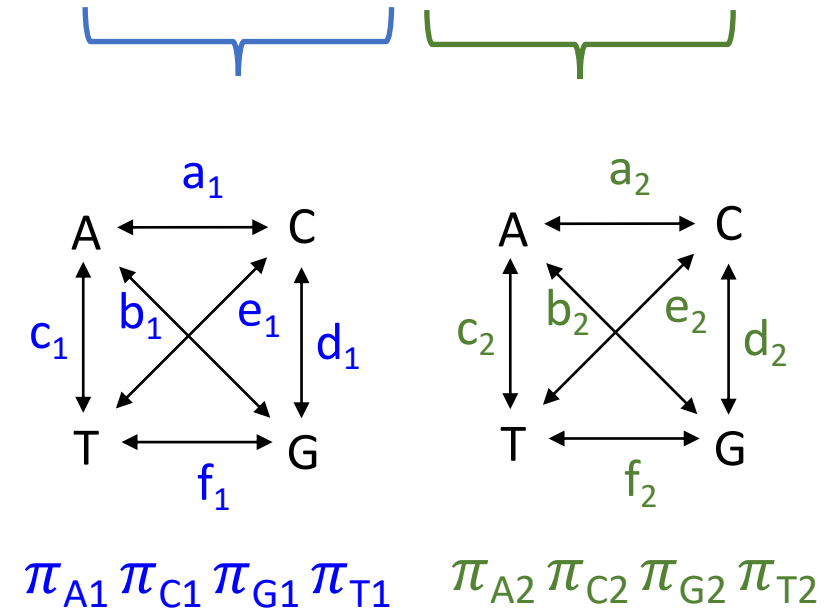
	Gene A	Gene B
Bird	C A A - -	- A A T A
Crocodile	C A C A -	T A C - T
Turtle	C A C T A	T A A G T
Human	C A C - -	- A C A A



Partition model

- Each partition has a separate model
- Often it can fit the data much better than a single model
- However,
 - Gene information is absent
 - Gene boundary is not accurate
 - For protein, partition according to protein domain?

	Gene A	Gene B
Bird	C A A - -	- A A T A
Crocodile	C A C A -	T A C - T
Turtle	C A C T A	T A A G T
Human	C A C - -	- A C A A



Mixture model

It **does not need** any partition information

Assume each site evolves under a mixture of ***k-class*** models

	Gene A				Gene B					
Bird	C	A	A	-	-	-	A	A	T	A
Crocodile	C	A	C	A	-	T	A	C	-	T
Turtle	C	A	C	T	A	T	A	A	G	T
Human	C	A	C	-	-	-	A	C	A	A

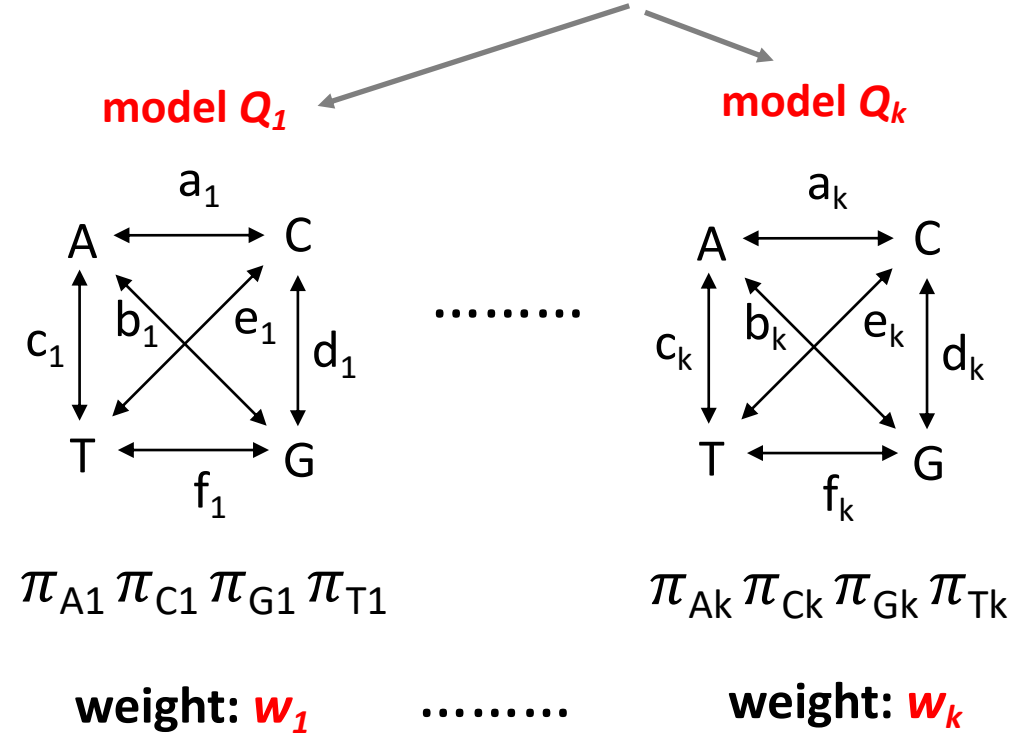
Each class can have a **different** model

For example: Q_1 – GTR; Q_2 – HKY; Q_3 – GTR'

Likelihood of each site is a **weighted sum** of likelihoods across all the models

For example, likelihood of site *i*:

$$L_i = w_1 L(Q_1 | D_i) + \dots + w_k L(Q_k | D_i)$$



Mixture model

Has been implemented in IQ-TREE for some years

However, this model is not popular

Not easy to decide

- the number of classes (i.e. the value of k)?
- which model for each class?

GTR

HKY

K80

	Gene A						Gene B			
Bird	C	A	A	-	-	-	A	A	T	A
Crocodile	C	A	C	A	-	T	A	C	-	T
Turtle	C	A	C	T	A	T	A	A	G	T
Human	C	A	C	-	-	-	A	C	A	A

model Q_1

model Q_k



Huaiyan Ren

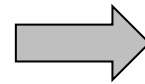
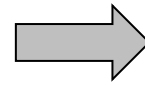
- Introduce a new algorithm **MixtureFinder**
- Automatically estimate
 - a. Optimal number of classes
 - b. Optimal model for each class

Is a single model good enough?

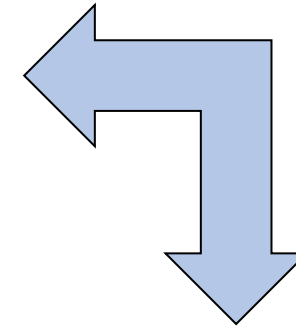
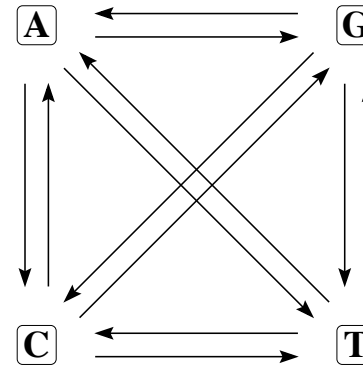
Examine 19,834 DNA partitions

Sequence Alignment

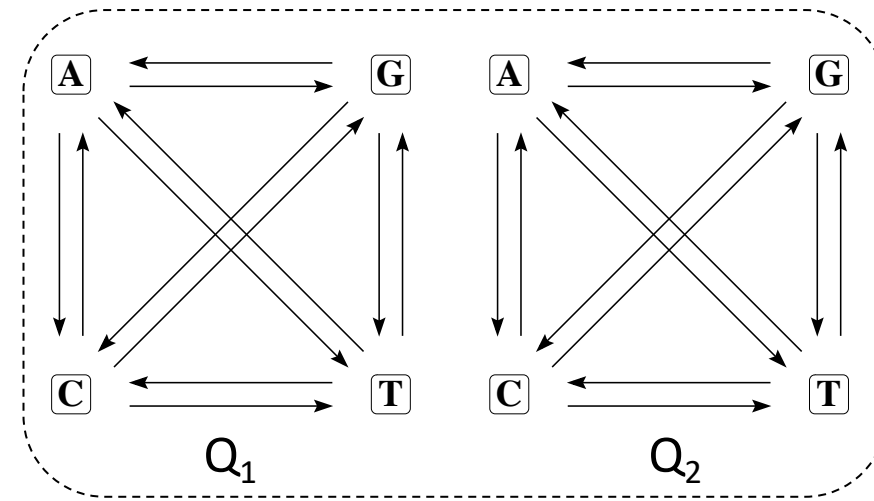
Bird	C	A	A	-	-	-	A	A	T	A
Crocodile	C	A	C	A	-	T	A	C	-	T
Turtle	C	A	C	T	A	T	A	A	G	T
Human	C	A	C	-	-	-	A	C	A	A



Markov model



Is mixture model better? (using BIC)

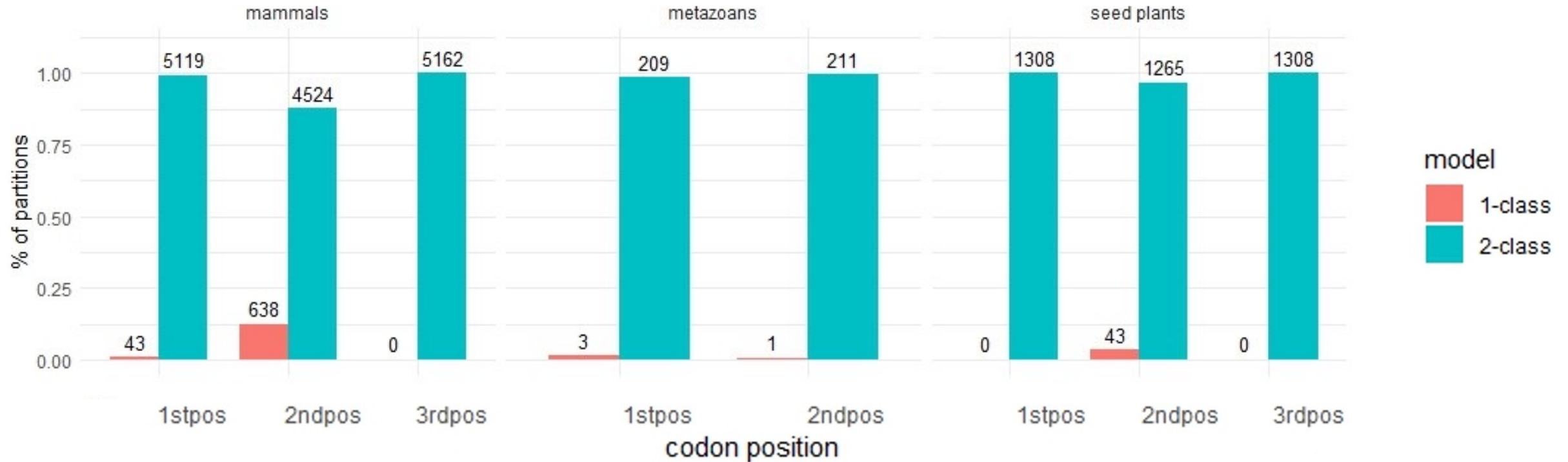


Mixture model of Q_1 and Q_2

Is one model enough?

Results

Among **19,834** DNA partitions, in **19,106 (96.3%)** of them the **2-class** models have **better** BIC value.



MixtureFinder (for DNA)

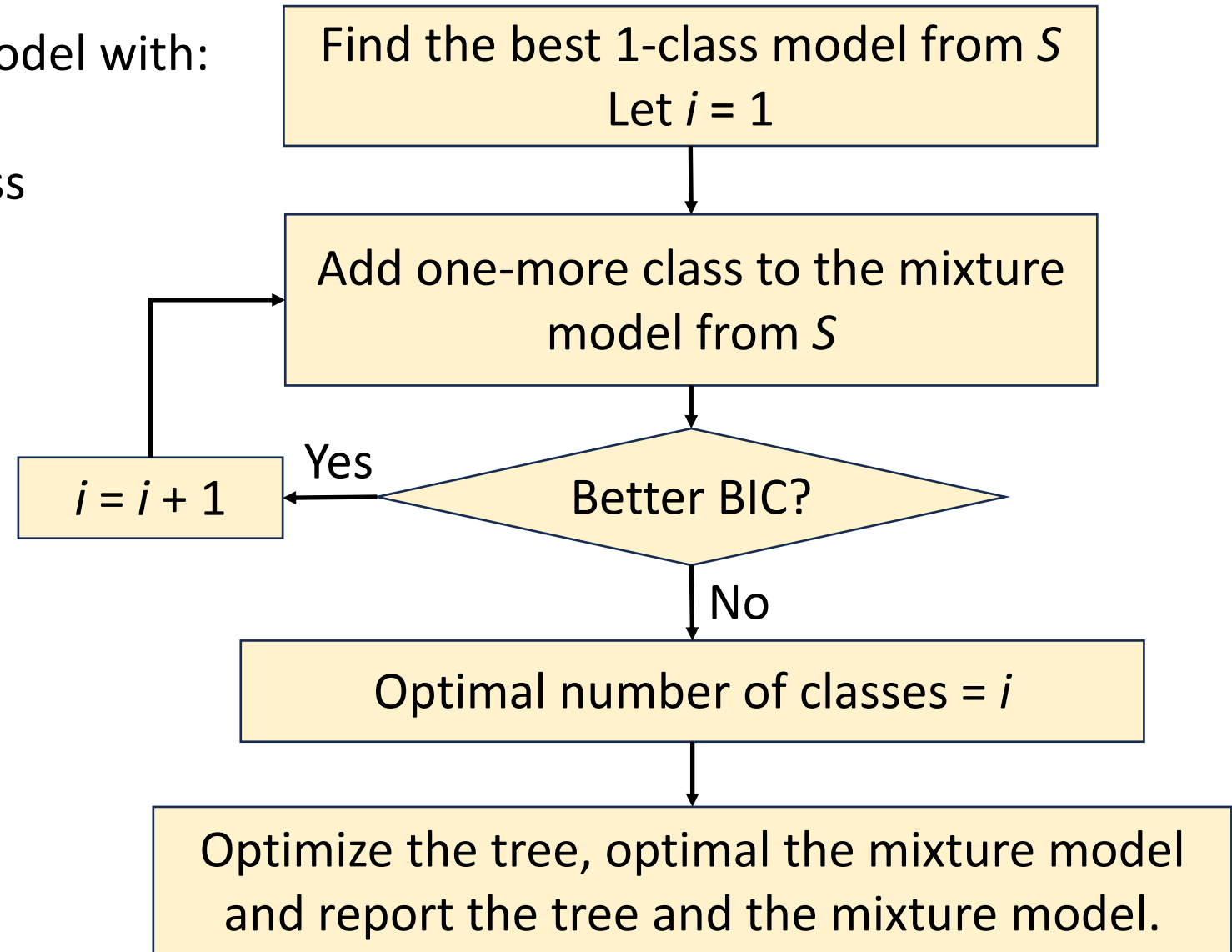
To compute the optimal mixture model with:

- Optimal number of classes
- Optimal model for each class

Let S be all DNA models to consider.

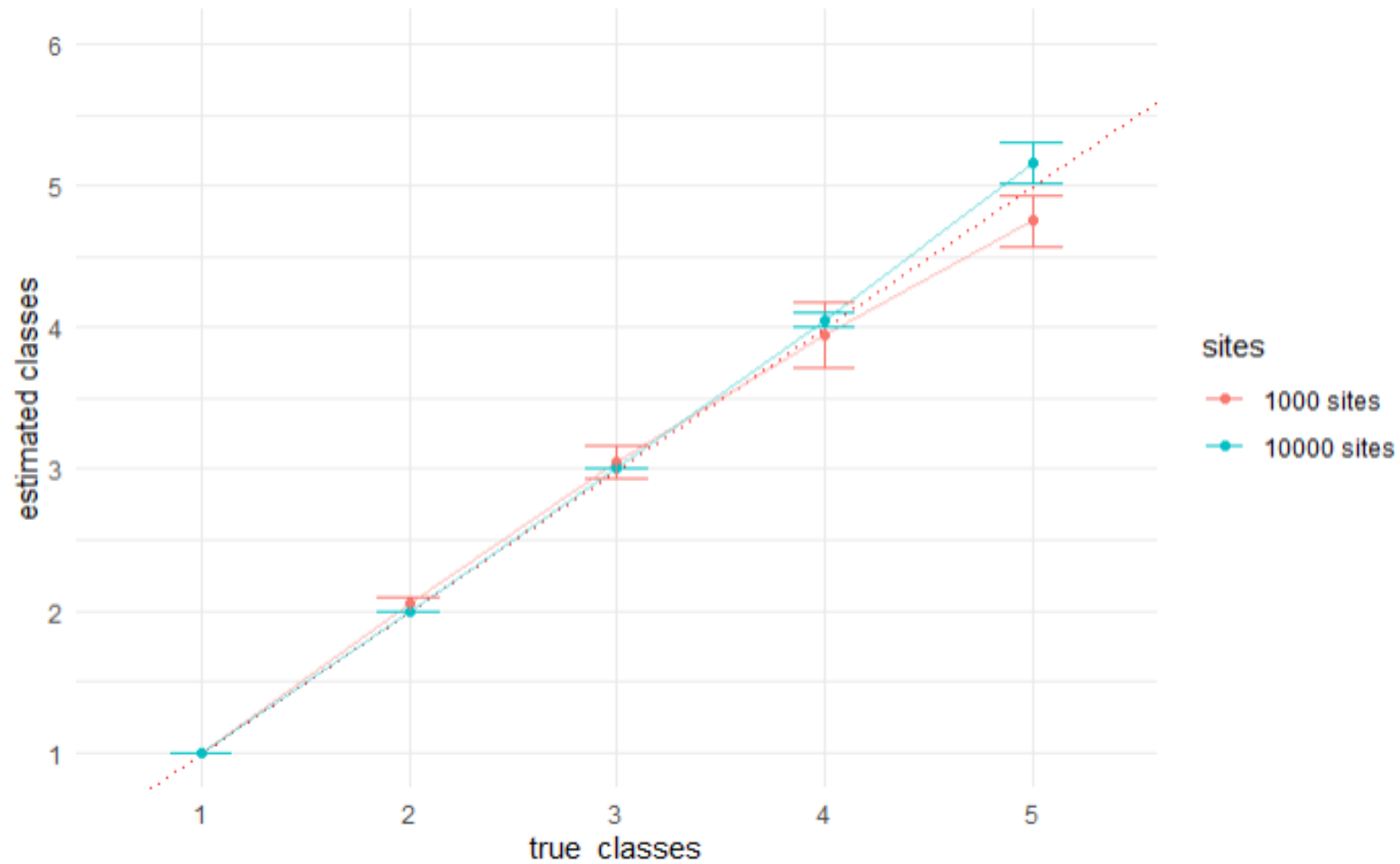
$S = \{JC, F81, K80, HKY, \dots, GTR\}$

Workflow



Evaluating MixtureFinder on Simulated data

Estimated number of classes vs simulated number of classes



Experiment 3

Apply the MixtureFinder on the empirical data:

[Home](#) > [BMC Biology](#) > [Article](#)

Phylogenomic analyses support the position of turtles as the sister group of birds and crocodiles (Archosauria)



Research article | [Open access](#) | [Published: 27 July 2012](#)

Volume 10, article number 65, (2012) | [Cite this article](#)

[Download PDF](#)   You have full access to this [open access](#) article

16 vertebrate taxa,
248 genes (**187k** sites)

Apply the MixtureFinder to
the **concatenated** alignment

[Ylenia Chiari](#) , [Vincent Cahais](#), [Nicolas Galtier](#) & [Frédéric Delsuc](#) 

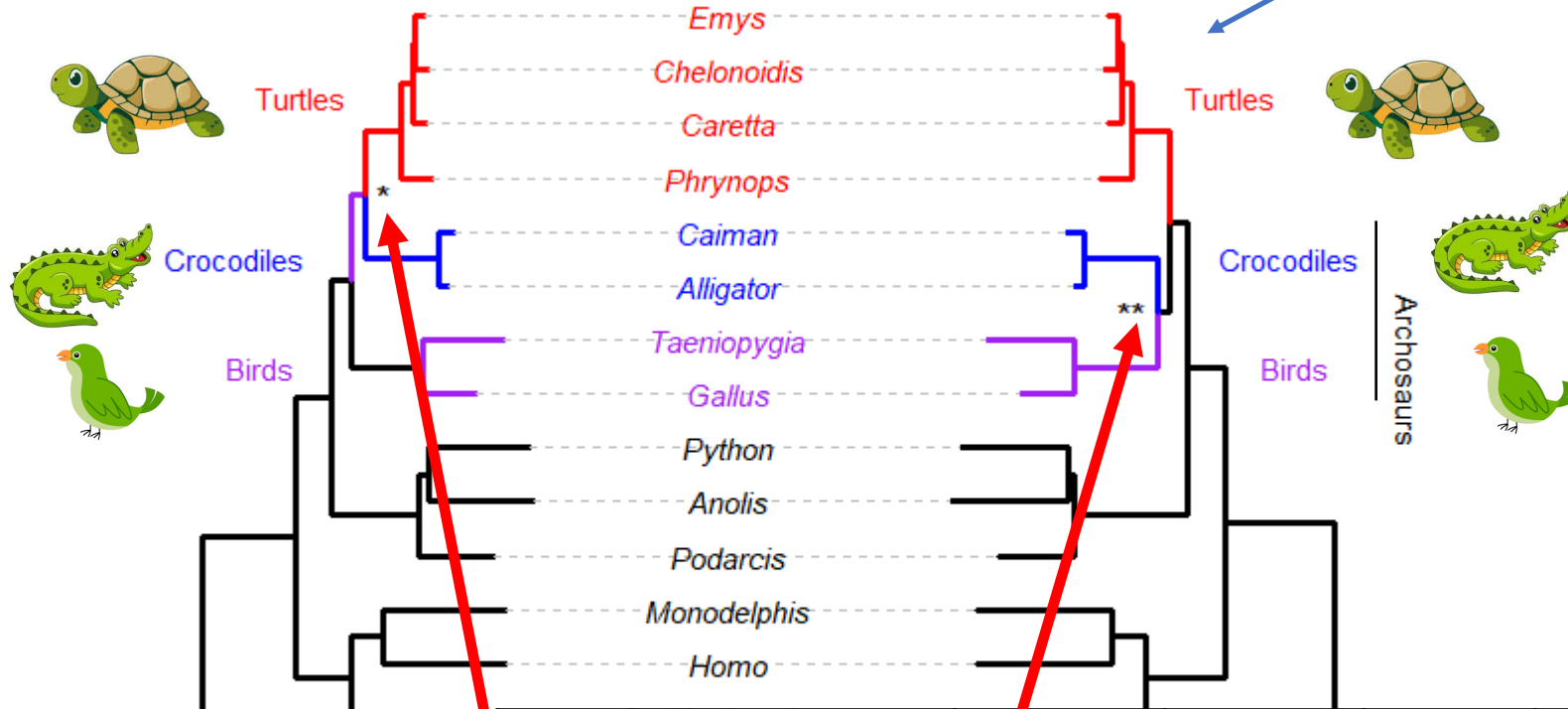
 37k Accesses  254 Citations  176 Altmetric  27 Mentions [Explore all metrics](#) →

Experiment 3

Apply the MixtureFinder on the empirical data:

This is the published tree

Single-class model tree



MixtureFinder tree

classes	1	2	3	4	5
topology	left	left	right	right	right
Bootstrap of node *	71	52	24	5	1
Bootstrap of node **	29	48	86	95	99

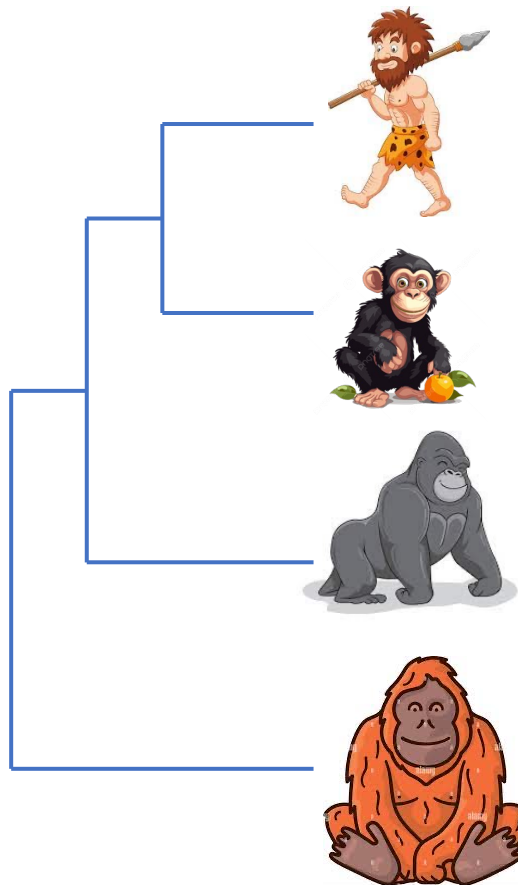
Conclusion

- A single substitution model is always **not adequate** to empirical data set.
- Model is **matter**! MixtureFinder can recover a better tree than a single model.
- MixtureFinder is **available in IQ-TREE2**.
- MixtureFinder now only supports **DNA**. We may extend it to amino acid in the future.

Preprint: <https://www.biorxiv.org/content/10.1101/2024.03.20.586035v2>

MAST model, a multi-tree mixture model

Traditional phylogenetic analyses always make this assumption:



C	A	A	-	-	-	A	A	T	A	T	T	A	C
C	A	C	A	-	T	A	C	-	T	T	T	A	C
C	A	C	T	A	T	A	A	G	T	T	T	A	C
C	A	C	-	-	-	A	C	A	A	A	T	A	C

Gene A

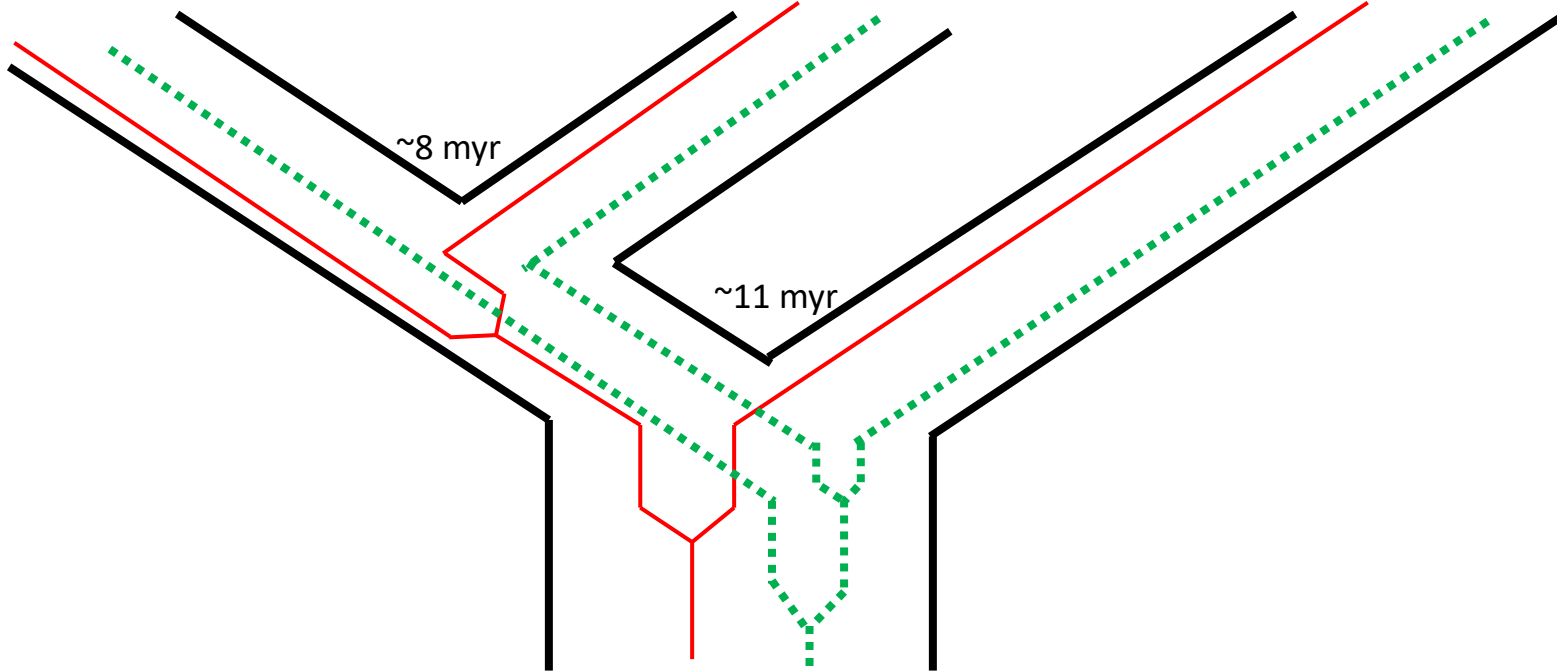
Gene B

Is it always true?

All sites evolved under a single tree

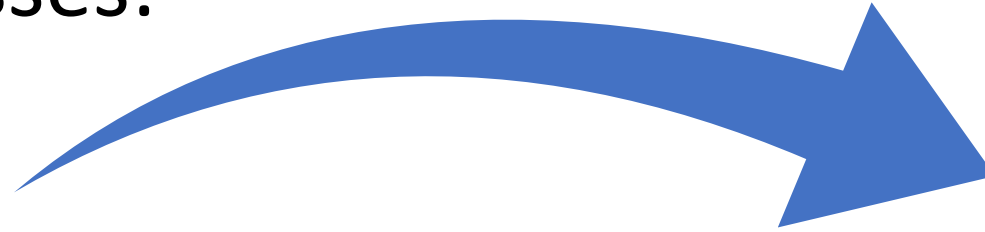
For example:

Incomplete lineage sorting



Biological processes:

Different genomic loci may have evolved under different trees



Incomplete lineage sorting

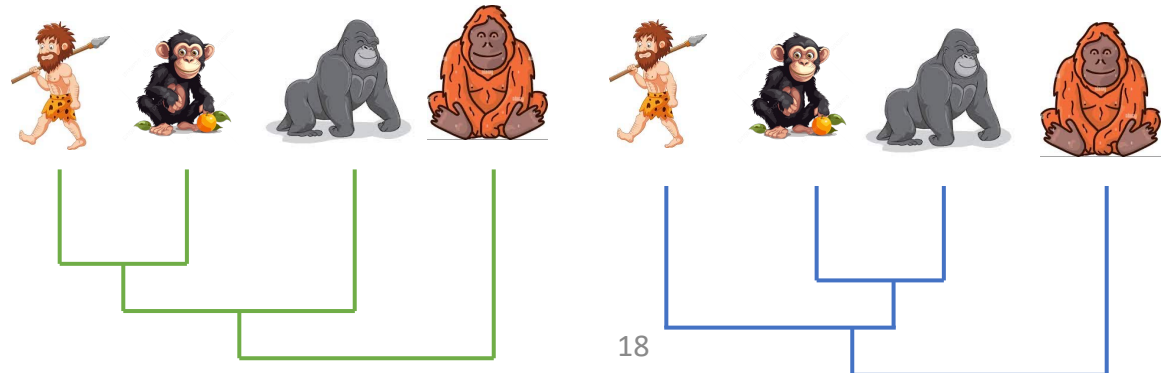
Introgression

Recombination

S1:	A	A	A	-	T	A	A	A	T	T	A	C
S2:	C	T	A	A	C	C	T	T	T	T	A	C
S3:	C	T	A	T	A	A	G	T	T	T	T	A
S4:	C	A	C	-	A	C	A	A	A	T	A	C

Gene A

Gene B



Existing approaches

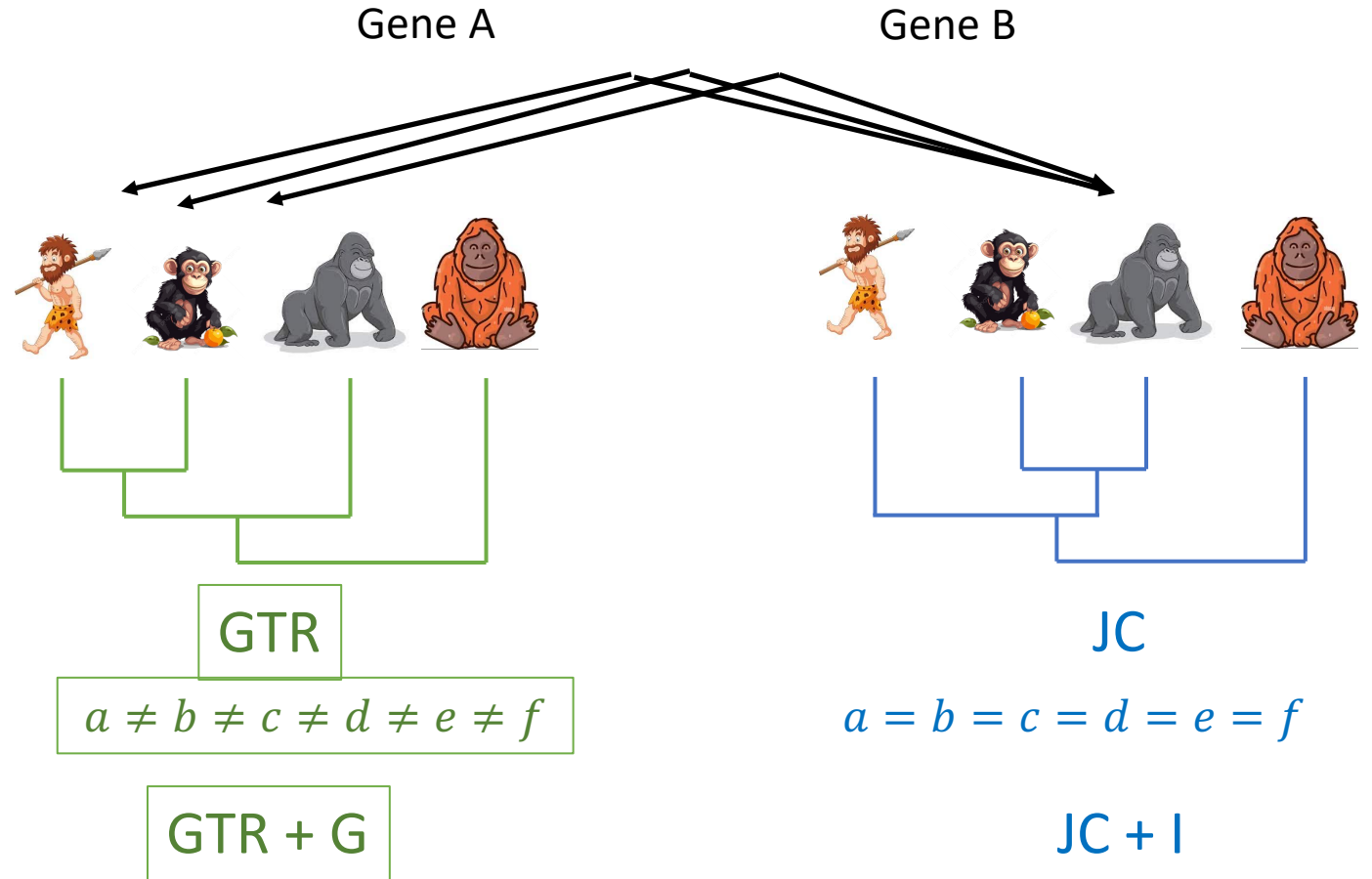
- Multi-species coalescent model
- Phylogenetic network

We suggest a new approach:
a Mixture Across Sites and Trees (MAST) model

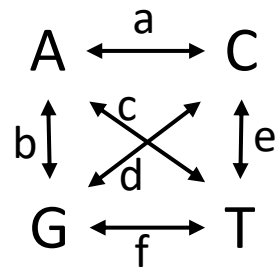
MAST model

S1:	A	A	A	-	T	A	A	A	T	T	A	C
S2:	C	T	A	A	C	C	T	T	T	T	A	C
S3:	C	T	A	T	A	A	G	T	T	T	T	A
S4:	C	A	C	-	A	C	A	A	A	T	A	C

Assume each site evolved under a mixture of trees



Substitution model



RHAS model

MAST model

Concatenated alignment

S1:	A	A	-	T	A	A	A	T
S2:	T	A	A	C	C	T	T	T
S3:	T	A	T	A	A	G	T	T
S4:	A	C	-	A	C	A	A	A

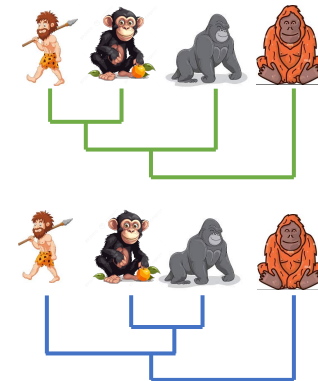
Gene A

Gene B

$L_{1,1}$ $L_{2,1}$ $L_{3,1}$ $L_{4,1}$ $L_{5,1}$ $L_{6,1}$ $L_{7,1}$ $L_{8,1}$

$L_{1,2}$ $L_{2,2}$ $L_{3,2}$ $L_{4,2}$ $L_{5,2}$ $L_{6,2}$ $L_{7,2}$ $L_{8,2}$

$$\text{Likelihood of site } i (L_i) = w_1 L_{i,1} + w_2 L_{i,2}$$

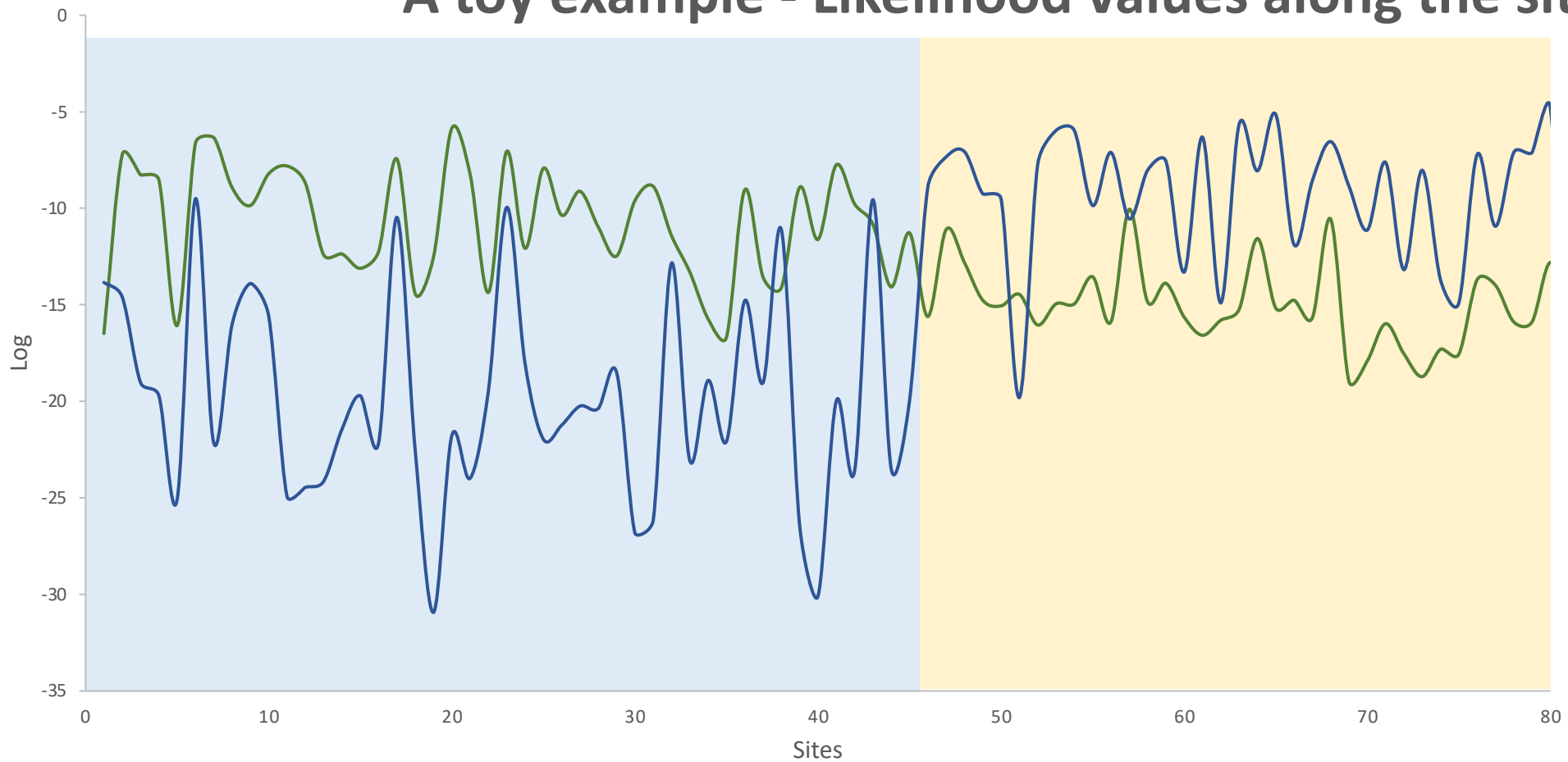


Weight of each tree

w_1

w_2

A toy example - Likelihood values along the sites

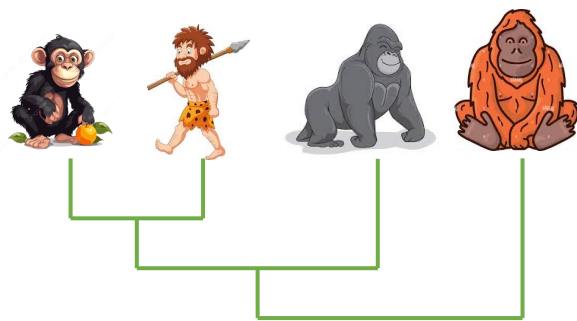


Likelihoods:

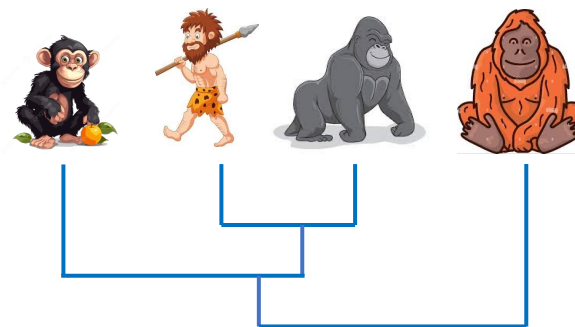
— Single-tree model (L_1)
for tree 1

— Single-tree model (L_2)
for tree 2

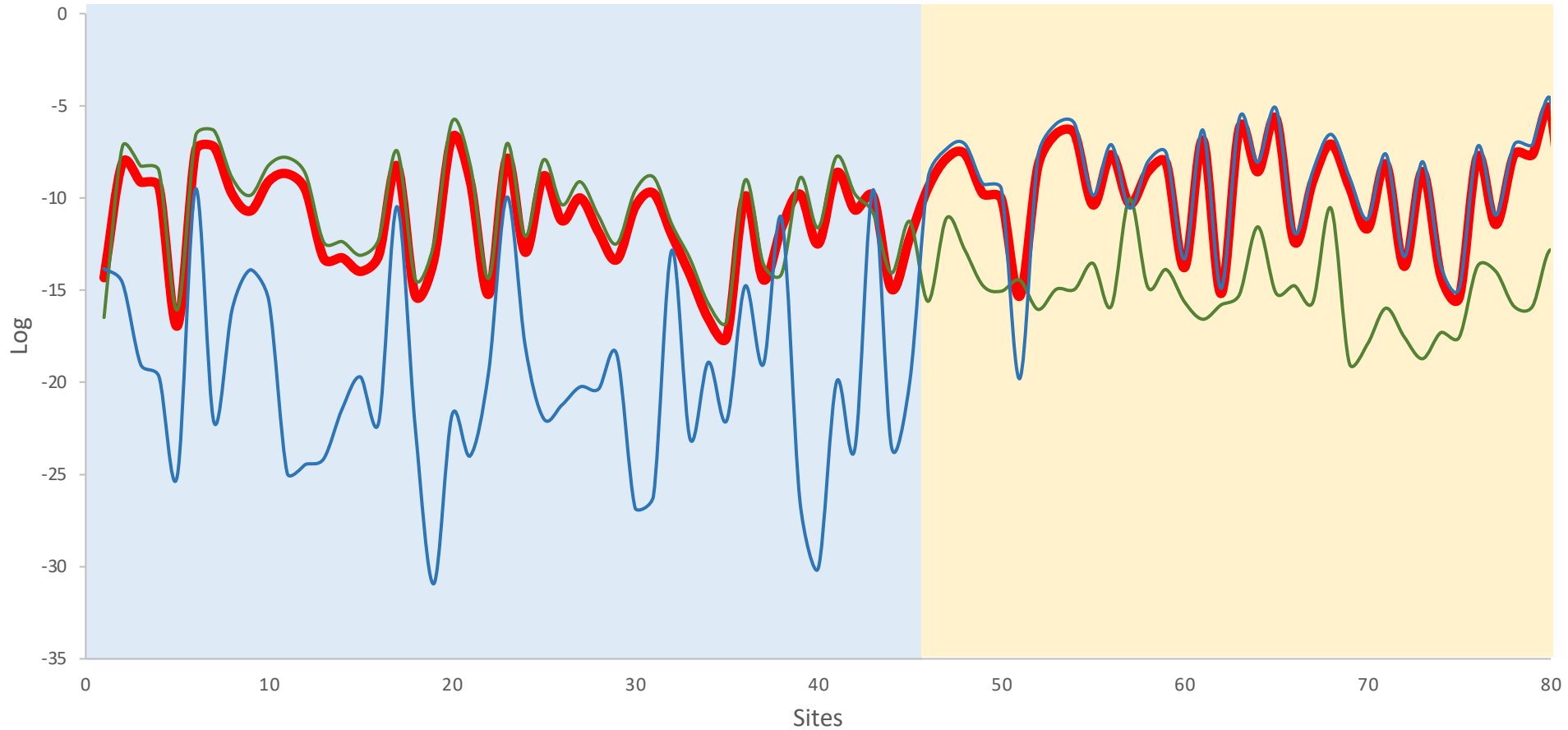
Tree 1



Tree 2



A toy example - Likelihood values along the sites



Likelihoods:

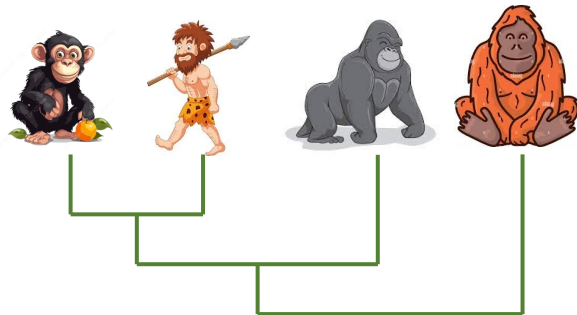
- MAST with both trees as an input (L_{MAST})
- Single-tree model for tree 1 (L_1)
- Single-tree model for tree 2 (L_2)

$$L_{MAST} = w_1 L_1 + w_2 L_2$$

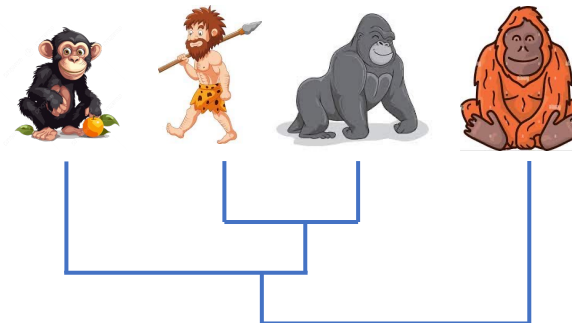
w_1 : weight of tree 1

w_2 : weight of tree 2

Tree 1



Tree 2



- The MAST model has been implemented in IQ-TREE
- We've done lots of simulations to verify its correctness.
- Tutorial: <http://www.iqtree.org/doc/Complex-Models#multitree-models>

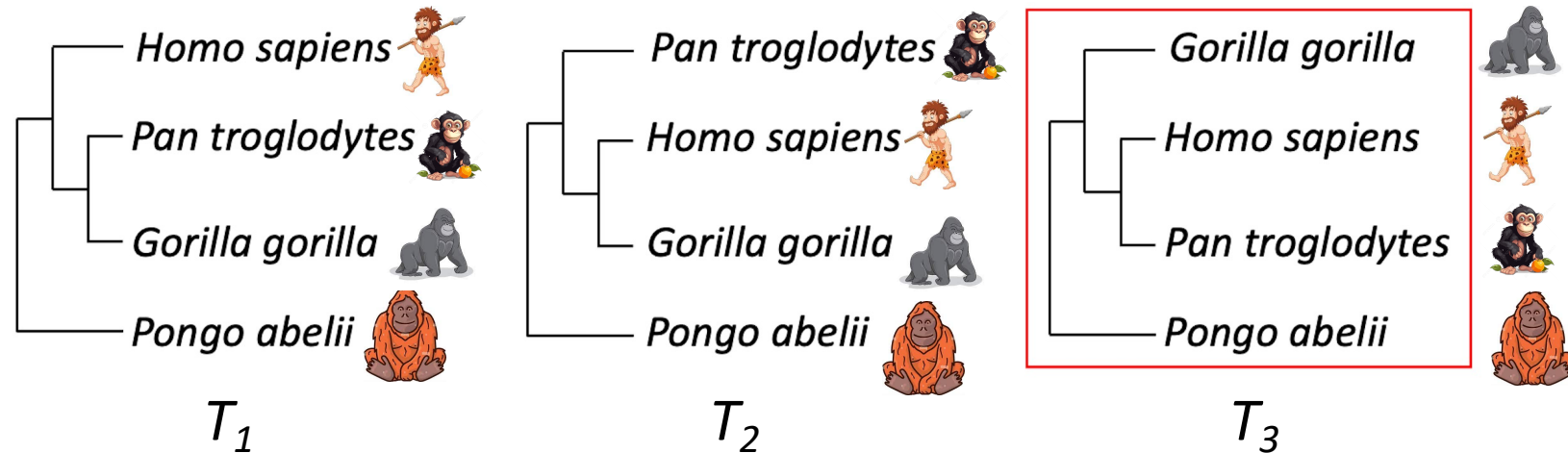
Three empirical experiments

Apply to Human-Chimp-Gorilla data

Number of genes: 1,595

Total length: 1,618,506

- Well-studied four-taxon grouping of human, chimpanzee, gorilla, and orangutan
- The accepted species tree: T_3



Gene tree frequencies:

19.8%

20.1%

60.1%

Our result

Best-fit MAST model weights:

17.9%

17.4%

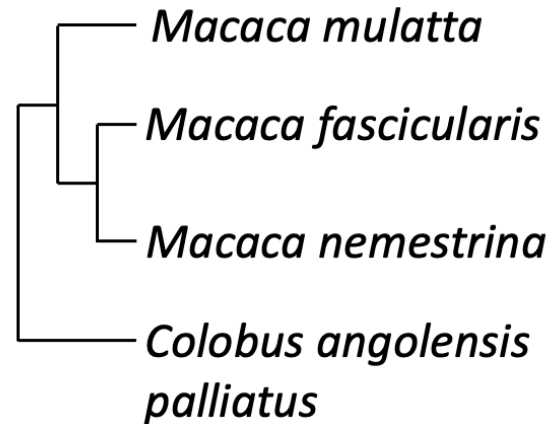
64.7%

- The frequencies of the minor trees from the MAST analysis are very similar
- Good indication of the existence of incomplete lineage sorting in this data set

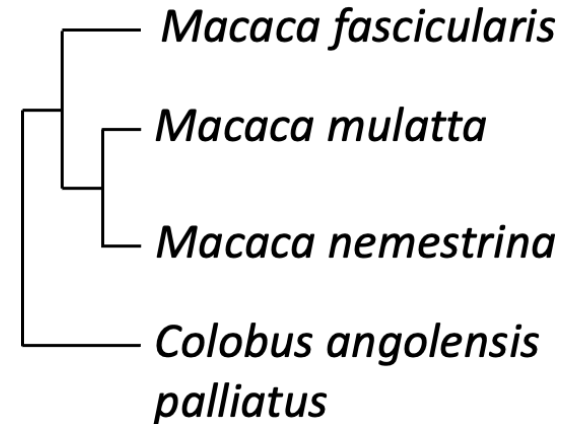
Number of genes: 1,599

Total length: 1,629,163

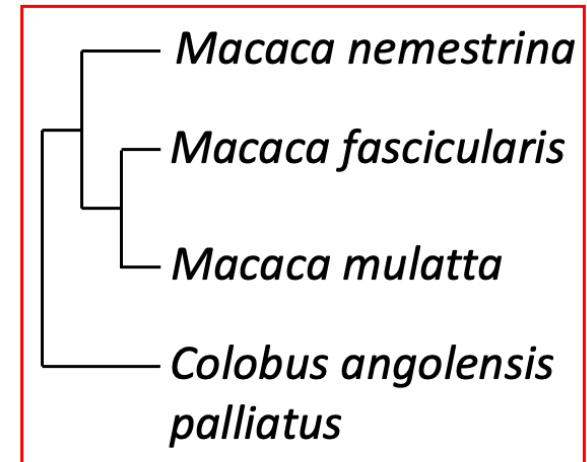
Apply to Macaques data



T_1



T_2



T_3

Gene tree frequencies:

31.2%

18.6%

50.2%

Best-fit MAST model weights:

17.3%

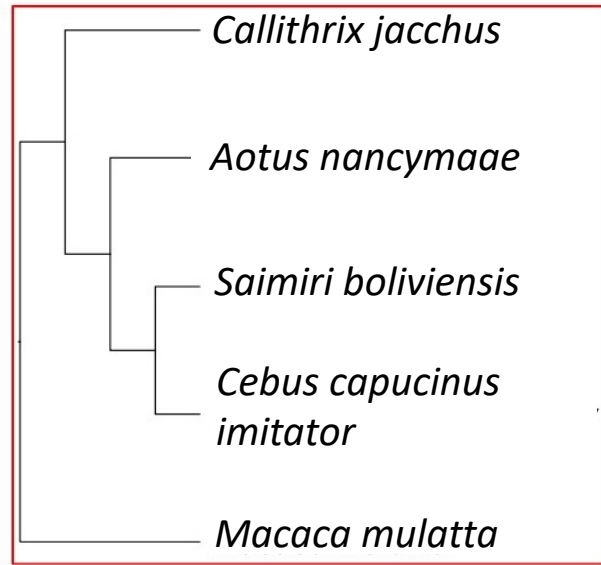
14.2%

68.5%

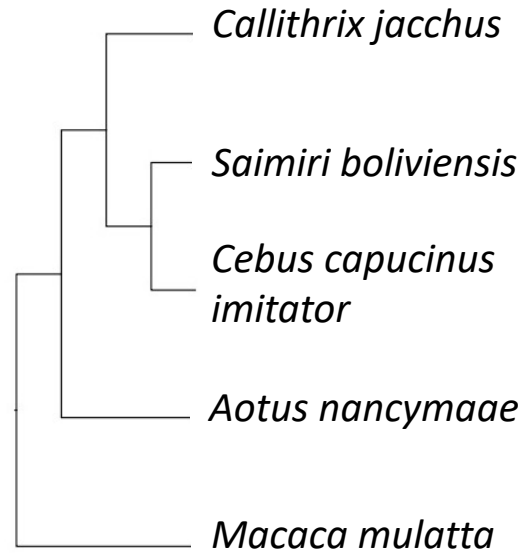
- The minor trees are substantially different in frequency from the MAST analysis
- Good indication of the existence of introgression in this data set

Apply to New World Monkeys

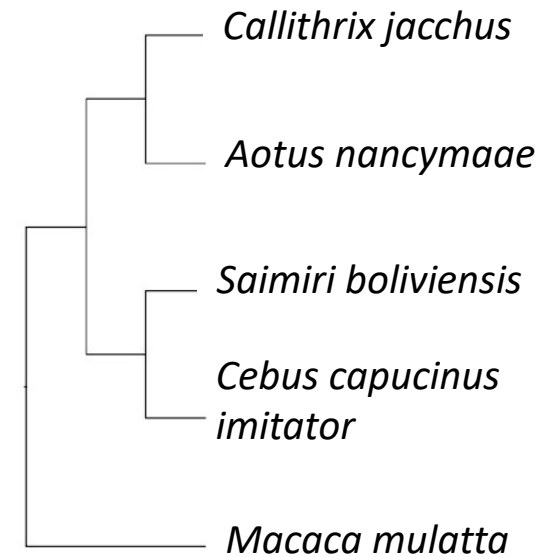
Number of genes: 1,557
Total length: 1,610,755



T_1



T_2



T_3

Gene tree frequencies:	37.1%	32.4%	30.5%
Single-tree model (IQ-TREE):			100%
Our result			
Best-fit MAST model weights:	42.4%	28.1%	29.6%

- The MAST model reported T_{D1} as the topology with the highest weight
- The MAST model can analyse a concatenated alignment using maximum likelihood, but without the biases that come with the single-tree assumption

Conclusion

1. The MAST model can overcome biases for a maximum likelihood method with single-tree assumption.
2. The weights estimated by the MAST model can be a good indication of some biological processes, like incomplete lineage sorting or introgression.

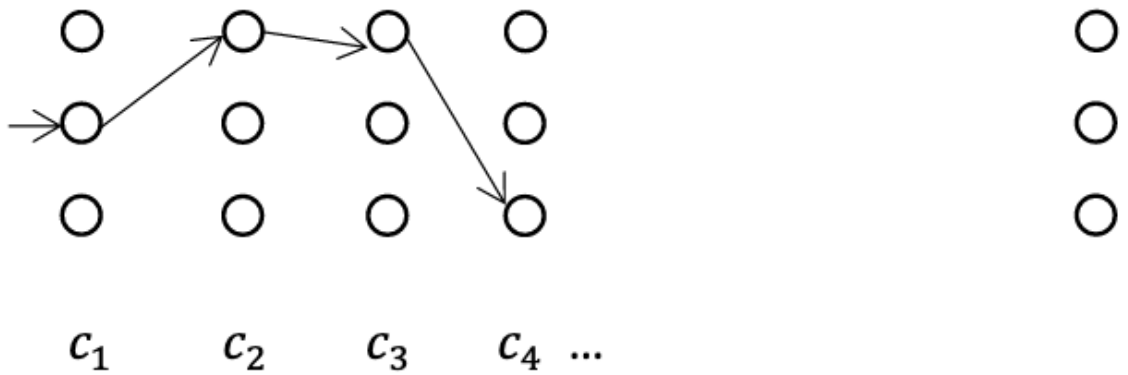
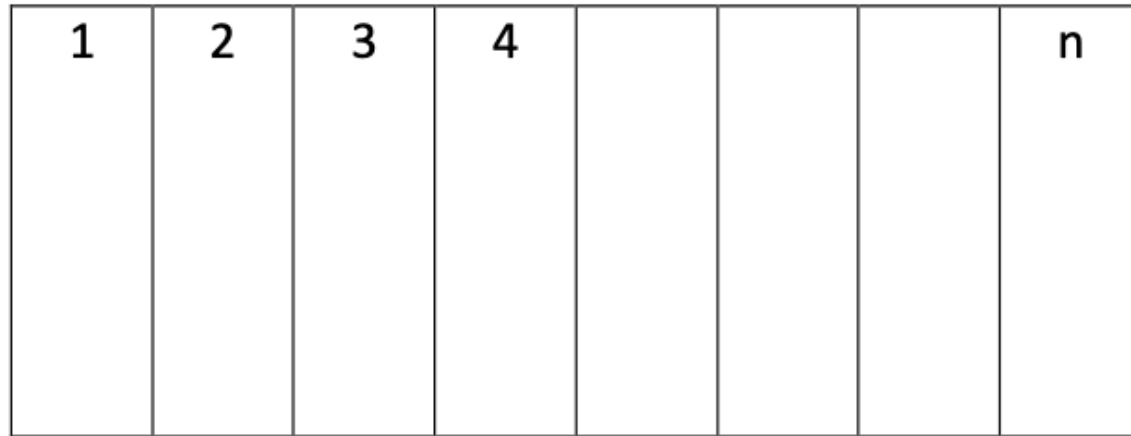
Published in Systematic Biology in March 2024
(<https://doi.org/10.1093/sysbio/syae008>)

HMM-MAST model, considering dependence between topologies along the adjacent sites (Preliminary results)

HMM-MAST: HMM on multi-tree model

- MAST model does not consider the dependence between the sites.
- Topologies along the sites should have dependence.
- The model is based on the paper:
J Felsenstein, G A Churchill, A Hidden Markov Model approach to variation among sites in rate of evolution., *Molecular Biology and Evolution*, Volume 13, Issue 1, Jan 1996, Pages 93–104, <https://doi.org/10.1093/oxfordjournals.molbev.a025575>

HMM-MAST: HMM on multi-tree model



- Assume each site has evolved under any of three trees $\{T_1, T_2, T_3\}$.
- However, this information is hidden.
- For each site, define the state c_i as the trees the site i may belong to.

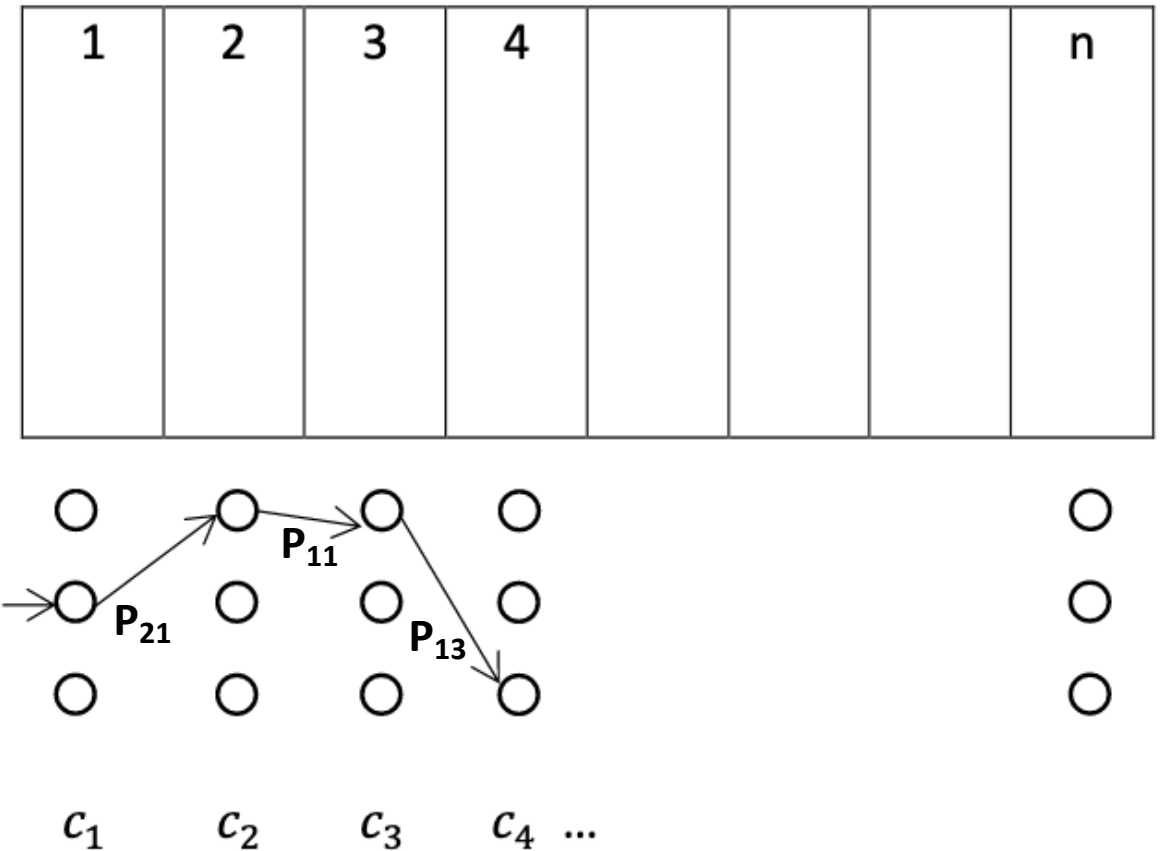
Parameters of the HMM-MAST

Transition probability matrix :

- Transition probability of going from c_{i-1} to c_i

$$P_{3 \times 3} = \begin{pmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{pmatrix}$$

The transition probability between the trees along the sites.



Backward probability formula

$$\Pr(D|T_1, T_2, T_3) = \sum_{c_1} \sum_{c_2} \dots \sum_{c_n} \Pr(c_1, c_2, \dots, c_n) \Pr(D|T_{c_1}, T_{c_2}, \dots, T_{c_n})$$

$$= \sum_{c_1=1}^k f_{c_1} \Pr\{D_1|T_{c_1}\} \sum_{c_2=1}^k p_{c_1 c_2} \Pr\{D_2|T_{c_2}\} \sum_{c_4=1}^k \dots \sum_{c_n=1}^k p_{c_{n-1} c_n} \Pr\{D_n|T_{c_n}\}$$

Difference between our method and PhyML-multi

- PhyML-multi also implemented HMM on phylogenetic.
- PhyML-multi assumes each site evolves under each topology with **equal** probability.
- PhyML-multi has difficulty in handling **long (> 5K) alignments**

Simulated data sets with partitions

- Various numbers of partitions with different numbers of sequences
- Each partition was simulated using a different tree
- The model used to simulate the data sets: GTR+G
- Every trees have different GTR and Gamma models
- For example:

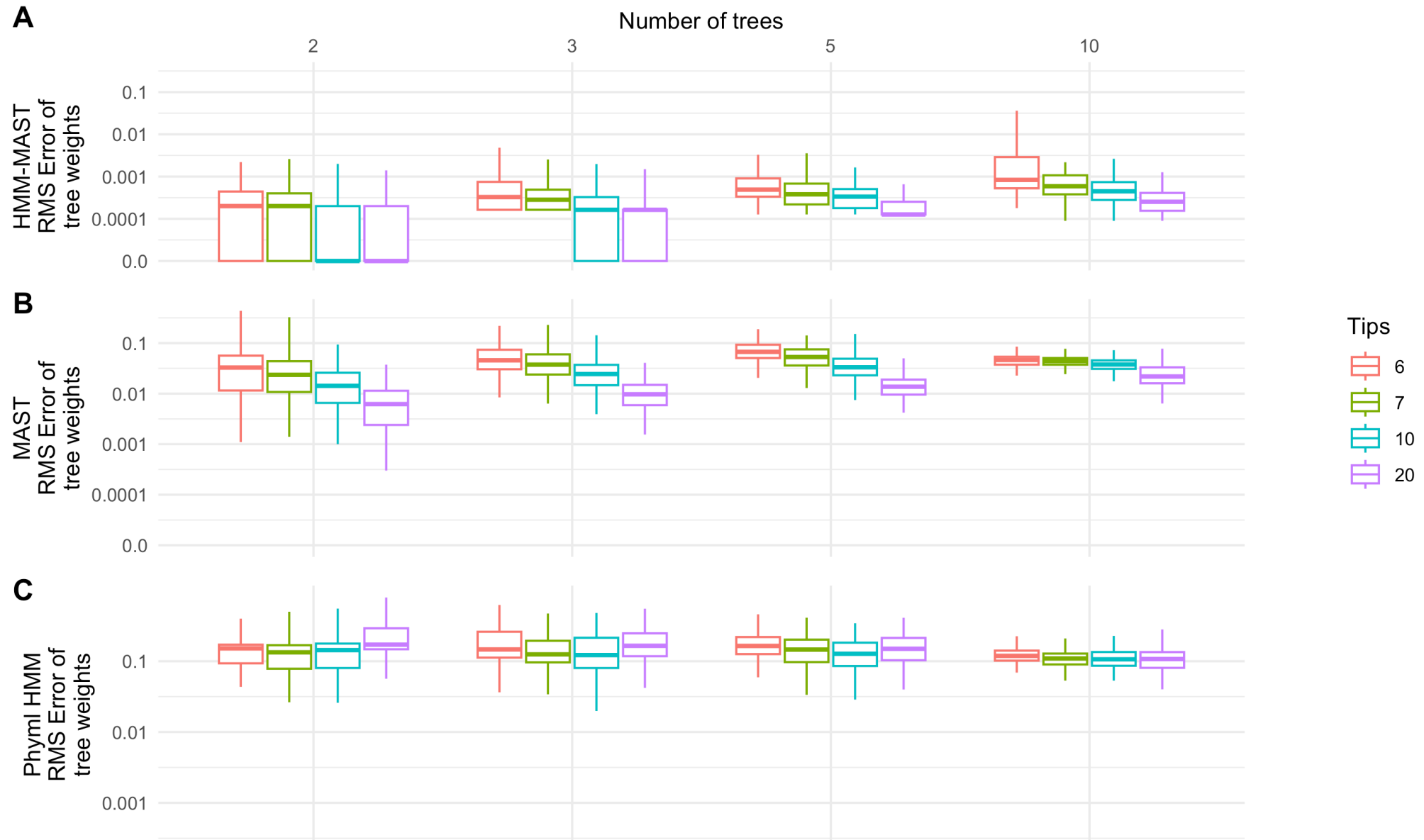


- Alignment length: 5K

Simulation results

Evaluate the proportion of sites belonging each tree

Alignment length: 5K

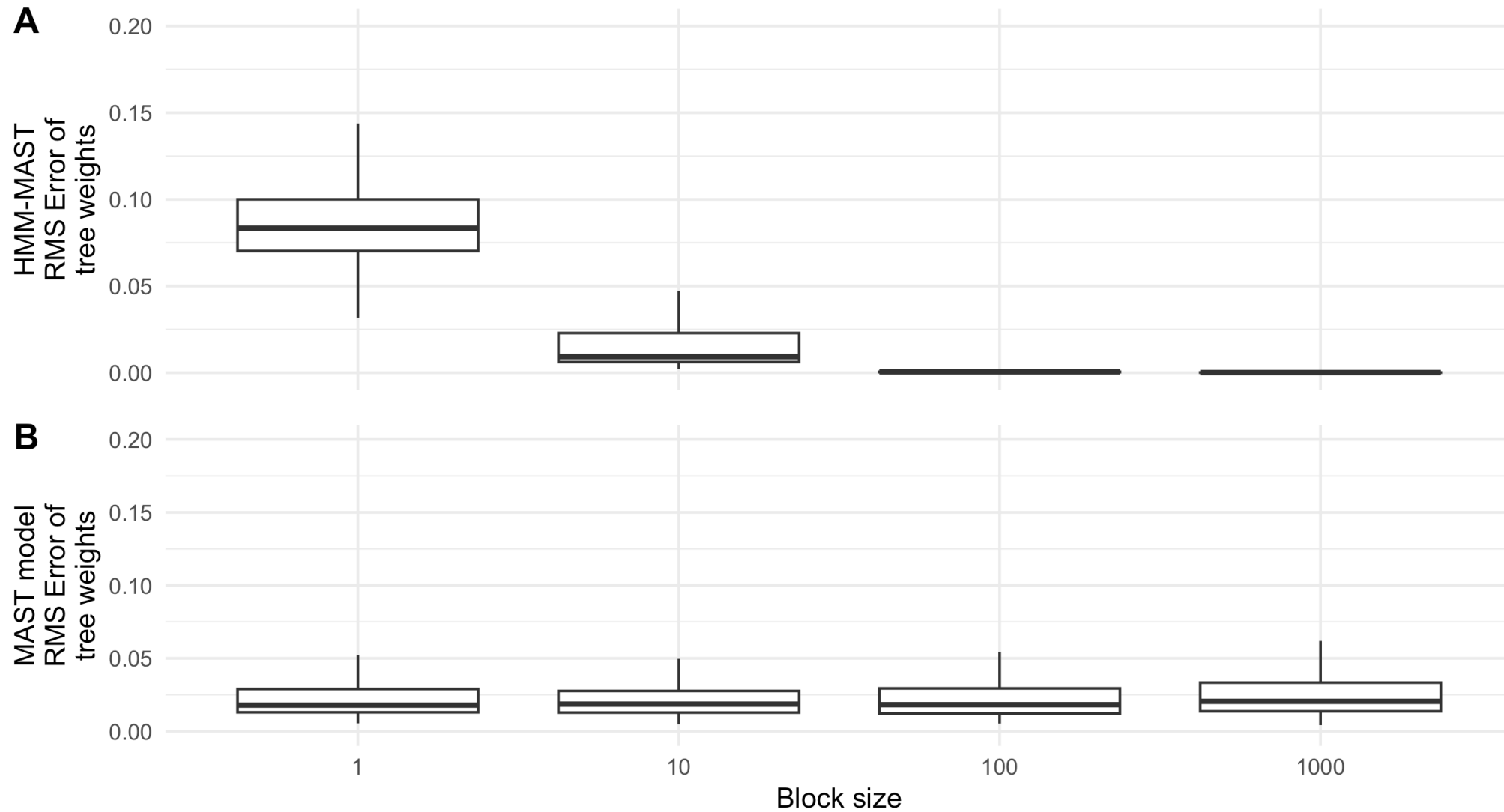


Simulated data sets with very short blocks

- Number of trees = 10
- Number of tips = 10
- Alignment length = 50K
- Block lengths: 1, 10, 100, 1000
- Each block was randomly assigned one tree
- The model used to simulate the data sets: GTR+G
- Every tree has different GTR and Gamma models
- For example (for 3 trees):



Results on simulated data sets with short blocks



Alignment lengths: 50K; Number of input trees = 10

Next direction.....

- HMM-MAST works well in simulated data sets
- Will perform testing on empirical data sets

Acknowledgements

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- Jeremias Ivan
- Nhan Trong Ly
- Rahil Vora

